EXHIBIT E

TO PLAINTIFFS' CLAIM CONSTRUCTION BRIEF PURSUANT TO P.R. 4-5(a)

Dot product

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In mathematics, the **dot product**, also known as the **scalar product**, is an operation which takes two vectors over the real numbers **R** and returns a real-valued scalar quantity. It is the standard **inner product** of the orthonormal Euclidean space. It contrasts with the cross product which produces a vector result.

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Definition

The dot product of two vectors $\mathbf{a} = [a_1, a_2, \dots, a_n]$ and $\mathbf{b} = [b_1, b_2, \dots, b_n]$ is defined as:

$$\mathbf{a} \cdot \mathbf{b} = \sum_{i=1}^{n} a_i b_i = a_1 b_1 + a_2 b_2 + \dots + a_n b_n$$

where Σ denotes summation notation and n is the dimension of the vectors.

For example, the dot product of two three-dimensional vectors [1, 3, -5] and [4, -2, -1] is

$$\begin{bmatrix} 1 & 3 & -5 \end{bmatrix} \cdot \begin{bmatrix} 4 & -2 & -1 \end{bmatrix} = (1)(4) + (3)(-2) + (-5)(-1) = 3.$$

For two complex column vectors the dot product is defined as

$$\mathbf{a} \cdot \mathbf{b} = \sum \overline{b_i} a_i$$

where $\overline{b_i}$ is the complex conjugate of b_i ; further note that in the complex case

$$\mathbf{a} \cdot \mathbf{b} = \overline{\mathbf{b} \cdot \mathbf{a}}$$

The dot product is typically applied to vectors from orthonormal vector spaces. Its generalization to non-orthonormal vector spaces is described below.

Conversion to matrix multiplication

Using matrix multiplication and treating the vectors as $n \times 1$ matrices (i.e. "column matrices" or "column vectors"), the dot product can also be written as:

$$\mathbf{a} \cdot \mathbf{b} = \mathbf{a}^T \mathbf{b}$$

where \mathbf{a}^{T} denotes the transpose of the matrix \mathbf{a} , and in this specific case, since \mathbf{a} is a column matrix, the transpose of \mathbf{a} is a "row matrix" or "row vector" ($1 \times n$ matrix).

For instance, the dot product of the two above-mentioned three-dimensional vectors is equivalent to the product of a 1×3 matrix by a 3×1 matrix (which, by virtue of the matrix multiplication, results in a 1×1 matrix, i.e., a scalar):

$$\begin{bmatrix} 1 & 3 & -5 \end{bmatrix} \begin{bmatrix} 4 \\ -2 \\ -1 \end{bmatrix} = \begin{bmatrix} 3 \end{bmatrix}.$$

Geometric interpretation

In Euclidean geometry, the dot product, length, and angle are related. For a vector **a**, the dot product **a** · **a** is the square of the length of **a**, or

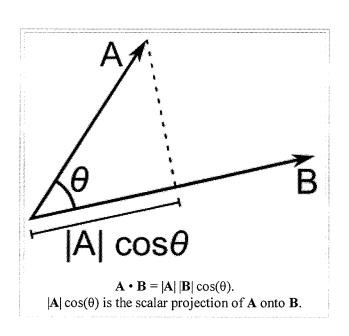
$$|\mathbf{a}| = \sqrt{\mathbf{a} \cdot \mathbf{a}}$$

where $|\mathbf{a}|$ denotes the length (magnitude) of \mathbf{a} . More generally, if \mathbf{b} is another vector

$$\mathbf{a} \cdot \mathbf{b} = |\mathbf{a}| |\mathbf{b}| \cos \theta$$

where $|\mathbf{a}|$ and $|\mathbf{b}|$ denote the length of \mathbf{a} and \mathbf{b} and θ is the angle between them.

Thus, given two vectors, the angle between them can be found by rearranging the above formula:



$$\theta = \arccos\left(\frac{\mathbf{a} \cdot \mathbf{b}}{|\mathbf{a}||\mathbf{b}|}\right).$$

As the cosine of 90° is zero, the dot product of two orthogonal vectors is always zero. Moreover, two vectors can be considered orthogonal if and only if their dot product is zero, and they have non-null length. This property provides a simple method to test the condition of orthogonality.

Sometimes these properties are also used for *defining* the dot product, especially in 2 and 3 dimensions; this definition is equivalent to the above one. For higher dimensions the formula can be used to define the concept of angle.

The geometric properties rely on the basis being orthonormal, i.e. composed of vectors perpendicular to each other and having unit length.

Scalar projection

If both a and b have length one (i.e. they are unit vectors), their dot product simply gives the cosine of the angle between them.

If only **b** is a unit vector, then the dot product $\mathbf{a} \cdot \mathbf{b}$ gives $|\mathbf{a}| \cos(\theta)$, i.e. the magnitude of the projection of **a** in the direction of **b**, with a minus sign if the direction is opposite. This is called the scalar projection of **a** onto **b**, or scalar component of **a** in the direction of **b** (see figure). This property of the dot product has several useful applications (for instance, see next section).

If neither **a** nor **b** is a unit vector, then the magnitude of the projection of **a** in the direction of **b**, for example, would be $\mathbf{a} \cdot (\mathbf{b} / |\mathbf{b}|)$ as the unit vector in the direction of **b** is $\mathbf{b} / |\mathbf{b}|$.

Rotation

A rotation of the orthonormal basis in terms of which vector \mathbf{a} is represented is obtained with a multiplication of \mathbf{a} by a rotation matrix \mathbf{R} . This matrix multiplication is just a compact representation of a sequence of dot products.

For instance, let

- $B_1 = \{\mathbf{x}, \mathbf{y}, \mathbf{z}\}$ and $B_2 = \{\mathbf{u}, \mathbf{v}, \mathbf{w}\}$ be two different orthonormal bases of the same space \mathbf{R}^3 , with B_2 obtained by just rotating B_1 ,
- $a_1 = (a_x, a_y, a_z)$ represent vector a in terms of B_1 ,
- $a_2 = (a_u, a_v, a_w)$ represent the same vector in terms of the rotated basis B_2 ,
- $\mathbf{u}_1, \mathbf{v}_1, \mathbf{w}_1$ be the rotated basis vectors $\mathbf{u}, \mathbf{v}, \mathbf{w}$ represented in terms of B_1 .

Then the rotation from B_1 to B_2 is performed as follows:

$$\mathbf{a}_2 = \mathbf{R}\mathbf{a}_1 = egin{bmatrix} u_x & u_y & u_z \ v_x & v_y & v_z \ w_x & w_y & w_z \end{bmatrix} egin{bmatrix} a_x \ a_y \ a_z \end{bmatrix} = egin{bmatrix} \mathbf{u}_1 \cdot \mathbf{a}_1 \ \mathbf{v}_1 \cdot \mathbf{a}_1 \ \mathbf{w}_1 \cdot \mathbf{a}_1 \end{bmatrix} = egin{bmatrix} a_u \ a_v \ a_w \end{bmatrix}.$$

Notice that the rotation matrix \mathbf{R} is assembled by using the rotated basis vectors \mathbf{u}_1 , \mathbf{v}_1 , \mathbf{w}_1 as its rows,

and these vectors are unit vectors. By definition, \mathbf{Ra}_1 consists of a sequence of dot products between each of the three rows of \mathbf{R} and vector \mathbf{a}_1 . Each of these dot products determines a scalar component of \mathbf{a} in the direction of a rotated basis vector (see previous section).

If a_1 is a row vector, rather than a column vector, then R must contain the rotated basis vectors in its columns, and must post-multiply a_1 :

$$\mathbf{a}_2 = \mathbf{a}_1 \mathbf{R} = \begin{bmatrix} a_x & a_y & a_z \end{bmatrix} \begin{bmatrix} u_x & v_x & w_x \\ u_y & v_y & w_y \\ u_z & v_z & w_z \end{bmatrix} = \begin{bmatrix} \mathbf{u}_1 \cdot \mathbf{a}_1 & \mathbf{v}_1 \cdot \mathbf{a}_1 & \mathbf{w}_1 \cdot \mathbf{a}_1 \end{bmatrix} = \begin{bmatrix} a_u \\ a_z & a_z \end{bmatrix}$$

Physics

In physics, magnitude is a scalar in the physical sense, i.e. a physical quantity independent of the coordinate system, expressed as the product of a numerical value and a physical unit, not just a number. The dot product is also a scalar in this sense, given by the formula, independent of the coordinate system. The formula in terms of coordinates is evaluated with not just numbers, but numbers times units. Therefore, although it relies on the basis being orthonormal, it does not depend on scaling.

Example:

Mechanical work is the dot product of force and displacement vectors.

Properties

The following properties hold if a, b, and c are real vectors and r is a scalar.

The dot product is commutative:

$$\mathbf{a} \cdot \mathbf{b} = \mathbf{b} \cdot \mathbf{a}$$

The dot product is distributive over vector addition:

$$\mathbf{a} \cdot (\mathbf{b} + \mathbf{c}) = \mathbf{a} \cdot \mathbf{b} + \mathbf{a} \cdot \mathbf{c}.$$

The dot product is bilinear:

$$\mathbf{a} \cdot (r\mathbf{b} + \mathbf{c}) = r(\mathbf{a} \cdot \mathbf{b}) + (\mathbf{a} \cdot \mathbf{c}).$$

When multiplied by a scalar value, dot product satisfies:

$$(c_1\mathbf{a})\cdot(c_2\mathbf{b})=(c_1c_2)(\mathbf{a}\cdot\mathbf{b})$$

(these last two properties follow from the first two).

Two non-zero vectors \mathbf{a} and \mathbf{b} are perpendicular if and only if $\mathbf{a} \cdot \mathbf{b} = 0$.

Unlike multiplication of ordinary numbers, where if ab = ac, then b always equals c unless a is zero, the

dot product does not obey the cancellation law:

If $\mathbf{a} \cdot \mathbf{b} = \mathbf{a} \cdot \mathbf{c}$ and $\mathbf{a} \neq \mathbf{0}$: then we can write: $\mathbf{a} \cdot (\mathbf{b} - \mathbf{c}) = 0$ by the distributive law; and from the previous result above: If \mathbf{a} is perpendicular to $(\mathbf{b} - \mathbf{c})$, we can have $(\mathbf{b} - \mathbf{c}) \neq \mathbf{0}$ and therefore $\mathbf{b} \neq \mathbf{c}$.

Provided that the basis is orthonormal, the dot product is invariant under isometric changes of the basis: rotations, reflections, and combinations, keeping the origin fixed. The above mentioned geometric interpretation relies on this property. In other words, for an orthonormal space with any number of dimensions, the dot product is invariant under a coordinate transformation based on an orthogonal matrix. This corresponds to the following two conditions:

- the new basis is again orthonormal (i.e., it is orthonormal expressed in the old one)
- the new base vectors have the same length as the old ones (i.e., unit length in terms of the old basis)

If **a** and **b** are functions, then the derivative of $\mathbf{a} \cdot \mathbf{b}$ is $\mathbf{a}' \cdot \mathbf{b} + \mathbf{a} \cdot \mathbf{b}'$

Triple product expansion

This is a very useful identity (also known as **Lagrange's formula**) involving the dot- and cross-products. It is written as

$$\mathbf{a} \times (\mathbf{b} \times \mathbf{c}) = \mathbf{b}(\mathbf{a} \cdot \mathbf{c}) - \mathbf{c}(\mathbf{a} \cdot \mathbf{b})$$

which is easier to remember as "BAC minus CAB", keeping in mind which vectors are dotted together. This formula is commonly used to simplify vector calculations in physics.

Proof of the geometric interpretation

Note: This proof is shown for 3-dimensional vectors, but is readily extendable to *n*-dimensional vectors.

Consider a vector

$$\mathbf{v} = v_1 \mathbf{i} + v_2 \mathbf{j} + v_3 \mathbf{k}.$$

Repeated application of the Pythagorean theorem yields for its length v

$$v^2 = v_1^2 + v_2^2 + v_3^2.$$

But this is the same as

$$\mathbf{v} \cdot \mathbf{v} = v_1^2 + v_2^2 + v_3^2$$

so we conclude that taking the dot product of a vector \mathbf{v} with itself yields the squared length of the vector.

Lemma 1

$$\mathbf{v} \cdot \mathbf{v} = v^2.$$

Now consider two vectors **a** and **b** extending from the origin, separated by an angle θ . A third vector **c** may be defined as

$$\mathbf{c} \stackrel{\text{def}}{=} \mathbf{a} - \mathbf{b}.$$

creating a triangle with sides a, b, and c. According to the law of cosines, we have

$$c^2 = a^2 + b^2 - 2ab\cos\theta.$$

Substituting dot products for the squared lengths according to Lemma 1, we get

$$\mathbf{c} \cdot \mathbf{c} = \mathbf{a} \cdot \mathbf{a} + \mathbf{b} \cdot \mathbf{b} - 2ab\cos\theta. \tag{1}$$

But as $\mathbf{c} \equiv \mathbf{a} - \mathbf{b}$, we also have

$$\mathbf{c} \cdot \mathbf{c} = (\mathbf{a} - \mathbf{b}) \cdot (\mathbf{a} - \mathbf{b}),$$

which, according to the distributive law, expands to

$$\mathbf{c} \cdot \mathbf{c} = \mathbf{a} \cdot \mathbf{a} + \mathbf{b} \cdot \mathbf{b} - 2(\mathbf{a} \cdot \mathbf{b}). \tag{2}$$

Merging the two $\mathbf{c} \cdot \mathbf{c}$ equations, (1) and (2), we obtain

$$\mathbf{a} \cdot \mathbf{a} + \mathbf{b} \cdot \mathbf{b} - 2(\mathbf{a} \cdot \mathbf{b}) = \mathbf{a} \cdot \mathbf{a} + \mathbf{b} \cdot \mathbf{b} - 2ab\cos\theta.$$

Subtracting $\mathbf{a} \cdot \mathbf{a} + \mathbf{b} \cdot \mathbf{b}$ from both sides and dividing by -2 leaves

$$\mathbf{a} \cdot \mathbf{b} = ab \cos \theta$$
.

Q.E.D.

Generalization

The inner product generalizes the dot product to abstract vector spaces and is normally denoted by < a, b>. Due to the geometric interpretation of the dot product the norm ||a|| of a vector a in such an inner product space is defined as

$$\|\mathbf{a}\| = \sqrt{\langle \mathbf{a}, \mathbf{a} \rangle}$$

such that it generalizes length, and the angle θ between two vectors \mathbf{a} and \mathbf{b} by

$$\cos \theta = \frac{\langle \mathbf{a}, \mathbf{b} \rangle}{\|\mathbf{a}\| \|\mathbf{b}\|}.$$

In particular, two vectors are considered orthogonal if their inner product is zero

$$\langle \mathbf{a}, \mathbf{b} \rangle = 0.$$

The Frobenius inner product generalizes the dot product to matrices. It is defined as the sum of the

products of the corresponding components of two matrices having the same size.

Matrix representation

An inner product can be represented using a square matrix and matrix multiplication. For example, given two vectors

$$\mathbf{a} = egin{bmatrix} a_u \ a_v \ a_w \end{bmatrix}, \qquad \mathbf{b} = egin{bmatrix} b_u \ b_v \ b_w \end{bmatrix}$$

with respect to the basis set S

$$S = \{\mathbf{u}, \mathbf{v}, \mathbf{w}\} = \left\{ \begin{bmatrix} u_1 \\ u_2 \\ u_3 \end{bmatrix}, \begin{bmatrix} v_1 \\ v_2 \\ v_3 \end{bmatrix}, \begin{bmatrix} w_1 \\ w_2 \\ w_3 \end{bmatrix} \right\}$$

any inner product can be represented as follows:

$$\langle \mathbf{a}, \mathbf{b} \rangle = \mathbf{a}^T \mathbf{M} \mathbf{b}$$

where M is a 3x3 matrix. Given the matrix of the inner products through S called C_S , M can be calculated by solving the following system of equations.

$$\mathbf{C}_{S} = \begin{bmatrix} \langle \mathbf{u}, \mathbf{u} \rangle & \langle \mathbf{u}, \mathbf{v} \rangle & \langle \mathbf{u}, \mathbf{w} \rangle \\ \langle \mathbf{v}, \mathbf{u} \rangle & \langle \mathbf{v}, \mathbf{v} \rangle & \langle \mathbf{v}, \mathbf{w} \rangle \\ \langle \mathbf{w}, \mathbf{u} \rangle & \langle \mathbf{w}, \mathbf{v} \rangle & \langle \mathbf{w}, \mathbf{w} \rangle \end{bmatrix} = \begin{bmatrix} \mathbf{u}^{T} \mathbf{M} \mathbf{u} & \mathbf{u}^{T} \mathbf{M} \mathbf{v} & \mathbf{u}^{T} \mathbf{M} \mathbf{w} \\ \mathbf{v}^{T} \mathbf{M} \mathbf{u} & \mathbf{v}^{T} \mathbf{M} \mathbf{v} & \mathbf{v}^{T} \mathbf{M} \mathbf{w} \\ \mathbf{w}^{T} \mathbf{M} \mathbf{u} & \mathbf{w}^{T} \mathbf{M} \mathbf{v} & \mathbf{w}^{T} \mathbf{M} \mathbf{w} \end{bmatrix}$$

If the basis set S is composed of orthogonal unit vectors (orthonormal basis), then both C_S and M reduce to the identity matrix 1, and the inner product can be represented as a simple product between a row matrix and a column matrix:

$$\langle \mathbf{a}, \mathbf{b} \rangle = \mathbf{a}^T \mathbf{1} \mathbf{b} = \mathbf{a}^T \mathbf{b}.$$

Thus, if the basis is orthonormal, the square matrix M is not necessary and the inner product coincides with the dot product as defined above:

$$\langle \mathbf{a}, \mathbf{b} \rangle = \mathbf{a}^T \mathbf{b} = \mathbf{a} \cdot \mathbf{b}.$$

Example

Given a basis set

$$S = \{\mathbf{u}, \mathbf{v}, \mathbf{w}\} = \left\{ \begin{bmatrix} 1\\0\\0 \end{bmatrix}, \begin{bmatrix} 1\\1\\0 \end{bmatrix}, \begin{bmatrix} 1\\1\\1 \end{bmatrix} \right\}$$

and a matrix of the inner product through S

$$\mathbf{C}_{S} = \begin{bmatrix} 5 & 2 & 0 \\ 2 & 6 & 2 \\ 0 & 2 & 7 \end{bmatrix}$$

we can set each element of C_S equal to the inner product of two of the basis vectors as follows

$$\mathbf{C}_S[i,j] = \langle \mathbf{S}[i], \mathbf{S}[j] \rangle$$

$$\mathbf{C}_{\mathcal{S}}[0,0] = 5 = \langle \mathbf{u}, \mathbf{u} \rangle = \begin{bmatrix} 1 & 0 & 0 \end{bmatrix} \mathbf{M} \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

$$\mathbf{C}_{S}[0,1] = 2 = \langle \mathbf{u}, \mathbf{v} \rangle = \begin{bmatrix} 1 & 0 & 0 \end{bmatrix} \mathbf{M} \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}$$

which gives nine equations and nine unknowns. Solving these equations yields

$$\mathbf{M} = \begin{bmatrix} 5 & -3 & -2 \\ -3 & 7 & -2 \\ -2 & -2 & 9 \end{bmatrix}.$$

Generalization to tensors

The dot product between a tensor of order n and a tensor of order m is a tensor of order n+m-2. The dot product is calculated by multiplying and summing across a single index in both tensors. If $\bf A$ and $\bf B$ are two tensors with element representation $A_{ij...}^{k\ell...}$ and $B_{mn...}^{p...i}$ the elements of the dot product $\bf A\cdot \bf B$ are given by

$$A_{ij...}^{k\ell...}B_{mn...}^{p...i} = \sum_{i=1}^{n} A_{ij...}^{k\ell...}B_{mn...}^{p...i}$$

This definition naturally reduces to the standard vector dot product when applied to vectors, and matrix multiplication when applied to matrices.

Occasionally, a double dot product is used to represent multiplying and summing across two indices. The double dot product between two 2nd order tensors is a scalar.

See also

- Physics
- Cross product
- Matrix multiplication
- Cauchy—Schwarz inequality

External links

- Weisstein, Eric W., "Dot product" from MathWorld.
- A quick geometrical derivation and interpretation of dot product (http://behindtheguesses.blogspot.com/2009/04/dot-and-cross-products.html)
- Interactive GeoGebra Applet (http://xahlee.org/SpecialPlaneCurves_dir/ggb/Vector Dot Product.html)
- Java demonstration of dot product (http://www.falstad.com/dotproduct/)
- Another Java demonstration of dot product (http://www.cs.brown.edu/exploratories/freeSoftware /repository/edu/brown/cs/exploratories/applets/dotProduct/dot_product_guide.html)
- Explanation of dot product including with complex vectors (http://www.mathreference.com/la,dot.html)
- "Dot Product" (http://demonstrations.wolfram.com/DotProduct/) by Bruce Torrence, Wolfram Demonstrations Project, 2007.

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Categories: Articles containing proofs | Bilinear forms | Linear algebra | Vectors | Analytic geometry

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